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# Use of TerraSAR-X Data to Retrieve Soil Moisture Over Bare Soil Agricultural Fields

Nicolas Baghdadi, Maelle Aubert, and Mehrez Zribi

**Abstract**—The retrieval of the bare soil moisture content from TerraSAR-X data is discussed using empirical approaches. Two cases were evaluated: 1) one image at low or high incidence angle and 2) two images, one at low incidence and one at high incidence. This study shows by using three databases collected between 2008 and 2010 over two study sites in France (Orgeval and Villamblain) that TerraSAR-X is a good remote sensing tool for the retrieving of surface soil moisture with accuracy of about 3% (rmse). Moreover, the accuracy of the soil moisture estimate does not improve when two incidence angles ( $26^\circ$ – $28^\circ$  or  $50^\circ$ – $52^\circ$ ) are used instead of only one. When compared with the result obtained with a high incidence angle ( $50^\circ$ – $52^\circ$ ), the use of low incidence angle ( $26^\circ$ – $28^\circ$ ) does not enable a significant improvement in estimating soil moisture (about 1%).

**Index Terms**—Soil moisture, TerraSAR-X images.

## I. INTRODUCTION

RADAR SIGNAL is a function of soil moisture and surface roughness in the case of bare soil. The possibility of retrieving these soil parameters was little investigated from X-band synthetic aperture radar (SAR). However, many studies were carried out by using C-band radar data (e.g., [1]–[4]). With the launch of satellites using the X-band ( $\sim 9.6$  GHz), such as TerraSAR-X and COSMO-SkyMed, the use of X-band data to derive soil parameters became possible. A radar configuration that minimizes the effects of surface roughness is recommended for a better estimate of soil moisture when using only one incidence angle. The optimal radar incidences in C-band for the retrieval of soil moisture are smaller than  $35^\circ$  [4].

Soil moisture estimation from SAR images is carried out by using physical or statistical models. Physical approach consists in using a physical model, such as the integral equation model [5], to predict the radar backscattering coefficient from SAR and soil parameters (wavelength, polarization, incidence angle, surface roughness, and soil dielectric constant). Statistical models based on experimental measurements are also often used in soil moisture estimation. For bare soils, the increase of radar angle ( $\sigma^\circ$ ) is supposed to be linear with the volumetric soil



Fig. 1. Location of study sites. (1) Orgeval. (2) Villamblain.

moisture for values between 5% and 35% [6]. Moreover,  $\sigma^\circ$  increases with soil surface roughness and follows an exponential or logarithmic behavior (e.g., [4] and [7]).

Very few studies analyzed the sensitivity of TerraSAR-X data to bare soil surface parameters. Baghdadi *et al.* [8] have observed that the radar signal at X-band is slightly more sensitive to surface roughness at high incidence angle than at low incidence angle. The difference observed between radar signals reflected by the roughest and smoothest areas increases with the radar wavelength. Moreover, results showed that the sensitivity of radar signal to surface roughness is better with PALSAR in L-band than with TerraSAR-X in X-band and that the C- and X-bands are similar sensitivity results. In this letter, only *in situ* soil moisture measurements in very wet conditions between 25% and 40% are available. Results obtained showed that the backscattering coefficient at X-band is stable when the moisture content ranges between 25% and 35% and that it decreases beyond this threshold.

Aubert *et al.* [9] have showed that the sensitivity of the TerraSAR-X signal to soil moisture is very important at low and high incidence angles. In comparison to results published with C-band SAR data, this sensitivity of the radar signal to soil moisture is higher in X-band. The second important result concerns the potential of the fine spatial resolution of TerraSAR (1 m) in the detection of soil moisture variations at the within-plot scale. The spatial distribution of slaking crust could be detected when soil moisture variation is observed between soil crusted and soil without crust. Indeed, areas covered by slaking crust could have greater soil moisture and, consequently, a greater backscattering signal than soils without crust.

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TABLE I  
 CHARACTERISTICS OF TERRASAR-X IMAGES AND SUMMARY OF GROUND-TRUTH MEASUREMENTS (*mv*, *rms*, AND *L*)

Date dd-mm-yy	Site	Pol.-Inc.	Fields number	<i>mv</i> (%) (min;max)	<i>rms</i> (cm) (min;max)	<i>L</i> (cm) (min;max)
06-02-08	Villamblain	HH-52°	8	(27 ; 34)	(1.3 ; 3.1)	(4.5 ; 9.1)
07-02-08	Villamblain	HH-28°	8	(27 ; 34)	(1.3 ; 3.1)	(4.5 ; 9.1)
12-02-08	Orgeval	HH-50°	6	(31 ; 36)	(1.8 ; 3.3)	(5.0 ; 9.3)
13-02-08	Orgeval	HH-26°	6	(31 ; 35)	(1.8 ; 3.3)	(5.0 ; 9.3)
17-03-09	Orgeval	HH-26°	7	(25 ; 32)	(1.7 ; 2.3)	(4.8 ; 6.9)
18-03-09	Orgeval	HH-50°	7	(24 ; 30)	(1.7 ; 2.3)	(4.8 ; 6.9)
25-03-09	Orgeval	HH-50°	3	(28 ; 29)	(2.0 ; 2.7)	(4.8 ; 5.7)
26-03-09	Orgeval	HH-26°	3	(24 ; 31)	(2.0 ; 2.7)	(4.8 ; 5.7)
08-04-09	Orgeval	HH-26°	6	(17 ; 26)	(1.1 ; 2.1)	(3.7 ; 6.0)
09-04-09	Orgeval	HH-50°	6	(15 ; 26)	(1.1 ; 2.1)	(3.7 ; 6.0)
01-03-10	Orgeval	HH-50°	6	(33 ; 40)	(1.9 ; 2.9)	(5.9 ; 7.5)
02-03-10	Orgeval	HH-26°	6	(33 ; 37)	(1.9 ; 2.9)	(5.9 ; 7.5)
12-03-10	Orgeval	HH-50°	7	(13 ; 25)	(1.1 ; 2.6)	(4.6 ; 7.0)
13-03-10	Orgeval	HH-26°	7	(15 ; 22)	(1.1 ; 2.6)	(4.6 ; 7.0)

71 At least one research question remained open. It concerns  
 72 the precision of the soil moisture estimates in bare agricultural  
 73 soils. The objective of this study is to examine the potential of  
 74 TerraSAR-X data for retrieving volumetric soil moisture over  
 75 bare soils. This work evaluates if the use of two incidence  
 76 angles at X-band [one low (26°–28°) and one high (50°–52°)]  
 77 improves the accuracy of the estimate of surface soil moisture  
 78 in comparison to only one incidence (low or high). TerraSAR-X  
 79 sensor has the advantage to acquire on the same study site  
 80 image pairs at low and high incidence angles within one day.  
 81 The goal of this work is to compare the findings with C- and  
 82 X-band data. At C-band, several studies have shown that the  
 83 use of two incidence angles provides distinct improvement in  
 84 the soil moisture estimate, in comparison with results obtained  
 85 using a single incidence (e.g., [1], [2], and [4]). Moreover,  
 86 low incidence angle is better than the high incidence angle  
 87 for estimating soil moisture with C-band SAR data. This letter  
 88 investigates this research question.

During the period of February–April (our SAR acquisitions), 99  
 the main crops are wheat and colza. They cover approximately 100  
 50% of the agricultural area. The remaining surface corre- 101  
 sponds to plowed soils awaiting future cultivation (corn and 102  
 potato). 103

### B. TerraSAR-X Images 104

Fourteen TerraSAR-X images (X-band ~9.65 GHz) were 105  
 acquired during the years of 2008, 2009, and 2010 (Table I). 106  
 The radar data are available in HH polarization, with incidence 107  
 angles ( $\theta$ ) of 26°, 28°, 50°, and 52°. The imaging mode 108  
 used was spotlight with a pixel spacing of 1 m. Radiometric 109  
 calibration using multilook ground range detected TerraSAR-X 110  
 images was first carried out using the following equation [10]: 111

$$\sigma_i^\circ(\text{dB}) = \log_{10}(Ks \cdot DN_i^2 - NEBN) + 10 \log_{10}(\sin \theta_i). \quad (1)$$

This equation transforms the amplitude of backscattered sig- 112  
 nal for each pixel ( $DN_i$ ) into a backscattering coefficient ( $\sigma^\circ$ ) 113  
 in decibels.  $Ks$  is the calibration coefficient, and  $NEBN$  is 114  
 the noise equivalent beta naught. All TerraSAR-X images were 115  
 then georeferenced using GPS points with a root-mean-square 116  
 error of the control points of approximately one pixel (i.e., 1 m). 117  
 This coregistration error was overcome by removing two 118  
 boundary pixels from each training plot relative to the limits 119  
 defined by the GPS control points. The mean backscattering 120  
 coefficients were calculated from calibrated SAR images by 121  
 averaging the linear  $\sigma^\circ$  values of all pixels within reference 122  
 fields. 123

## II. STUDY AREA AND DATA SET

89

### A. Study Site

91 Data were acquired over two mainly agricultural sites  
 92 (Fig. 1). The Villamblain site is located in the south of Paris,  
 93 France (latitude 48°01' N and longitude 1°35' E) with soil  
 94 composed of 30% clay, 60% silt, and 10% sand. The second  
 95 site is situated in the Orgeval watershed, located in the east of  
 96 Paris, France (latitude 48°51' N and longitude 3°07' E). The soil  
 97 has a loamy texture, composed of 78% silt, 17% clay, and 5%  
 98 sand. Both of these two sites are very flat.

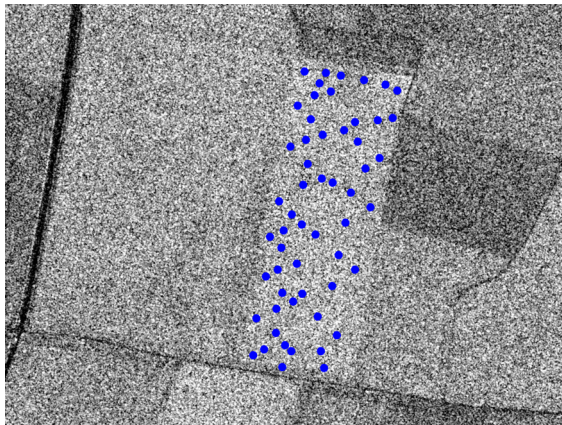


Fig. 2. Example of volumetric soil moisture measurements taken on a reference field.

### 124 C. Field Data

125 Simultaneously with TerraSAR-X acquisition, field mea-  
 126 surements of soil moisture and surface roughness have been  
 127 achieved on several bare soil reference fields of at least 2 ha.  
 128 In the case of TerraSAR-X in spotlight mode (pixel spacing of  
 129 1 m), this corresponds to a surface of 20 000 pixels or more.

130 The volumetric water content at field scale was assumed to be  
 131 equal to the mean value estimated from several samples (20–40  
 132 measurements per field; Fig. 2) collected from the top 5 cm  
 133 of soil using the gravimetric method. The soil moistures range  
 134 from 13% to 40%.

135 In most studies of microwave measurements carried out over  
 136 bare soils, the experimental relationship between soil moisture  
 137 and backscattering coefficient is provided by mean volumetric  
 138 water contents measured to a soil depth, generally 0–5 cm  
 139 or 0–10 cm. Indeed, only some studies using theory results  
 140 are available at X-band. These studies suggest a penetration  
 141 depth maybe lower than 5 cm. No experimental measurements  
 142 are made in field condition, and the low penetration depth  
 143 of X-band is only based on theoretical study. Therefore, the  
 144 penetration depth of the X-band is not yet well known.

145 Roughness measurements were made using needle pro-  
 146 filometers (1 m long and with 2-cm sampling intervals). Ten  
 147 roughness profiles were sampled for each training field (parallel  
 148 and perpendicular to the row direction). From these measure-  
 149 ments, the two roughness parameters, i.e., root mean square  
 150 (*rms*) surface height and correlation length (*L*), were calcu-  
 151 lated using the mean of all correlation functions. The *rms*  
 152 surface heights range from 1.1 to 3.3 cm, and the correlation  
 153 length (*L*) varies from 2.3 cm in sown fields to 9.3 cm in plowed  
 154 fields.

### 155 III. METHODOLOGY

156 The retrieval of soil moisture from TerraSAR-X images  
 157 by means of empirical approaches requires the development  
 158 of experimental relationships between  $\sigma_{\text{TerraSAR-X}}^{\circ}$  and the  
 159 measured soil moisture. TerraSAR data acquired in two config-  
 160 urations of incidence angles ( $\sim 26^{\circ}$  and  $\sim 50^{\circ}$ ) were used with  
 161 ground measurements conducted over bare soil. The sensitivity  
 162 of TerraSAR signal to soil moisture is the greatest for low

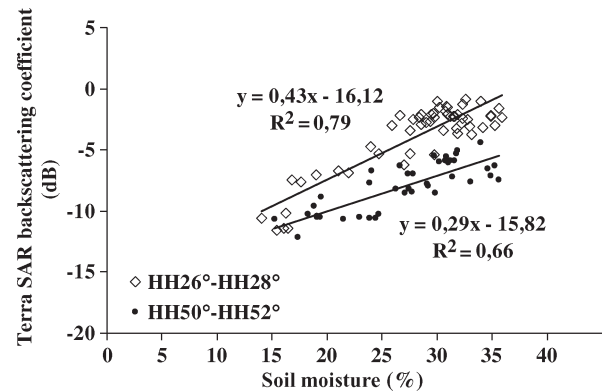


Fig. 3. TerraSAR-X signal versus volumetric soil moisture (measured at a depth of 5 cm). Each point corresponds to the average backscattering coefficient in decibels for one reference field. Thirty points are used for each of the two configurations HH26°–28° and HH50°–52° (data sets of 2008 and 2009).

incidence angle (0.43 dB/% for 26°–28° and 0.29 dB/% for 163  
 50°–52°; Fig. 3). For a confidence level of 95%, there are sig- 164  
 nificant relationships between the TerraSAR-X backscattering 165  
 coefficient and the *in situ* soil moisture because the *p*-values are 166  
 much less than 0.05 (*p*-value  $< 2.2 \times 10^{-16}$  for HH26°–28° 167  
 and *p*-value  $= 1.52 \times 10^{-10}$  for HH50°–52°). 168

169 Studies using C-band (ERS, RADARSAT, ASAR, etc.) 169  
 showed lower sensitivities between radar signal and soil mois- 170  
 ture, between 0.2 and 0.3 dB/% for low incidence angles 171  
 and about 0.1 dB/% for high incidence angles (e.g., [2] and 172  
 [11]–[13]). 173

174 The objective of this study is to analyze the influence of 174  
 incidence angle on the accuracy of the soil moisture estimate. 175  
 Configurations in HH polarization with single incidence an- 176  
 gle (26°–28° or 50°–52°) were studied. Next, multi-incidence 177  
 TerraSAR-X images acquired at both low and high  $\theta$  values 178  
 with one-day-spaced dates and only minor variations in soil 179  
 characteristics were used to analyze the possible improvement 180  
 in the soil moisture estimates when two incidences are used. 181

182 The empirical relationship between the radar backscattering 182  
 coefficient ( $\sigma^{\circ}$ ) and the volumetric soil moisture (*mv*) for bare 183  
 soil surfaces without taking into account the *rms* surface height 184  
 is given by (e.g., [14]; Fig. 3) 185

$$\sigma_{\text{dB}}^{\circ} = f(mv, \theta)_{\text{dB}} = \delta mv + \xi. \quad (2)$$

186 This simplified relationship is valid for *mv* values between 186  
 5% and 35% [6]. The coefficient  $\delta$  is dependent on SAR pa- 187  
 rameters (radar wavelength, incidence angle, and polarization), 188  
 while the coefficient  $\xi$  is controlled by SAR parameters and 189  
 surface roughness. Experimental data of  $\sigma^{\circ}$  and *mv* show slope 190  
 $\delta$  values of about 0.43 dB/% for HH26°–28° and 0.29 dB/% for 191  
 HH50°–52°. 192

193 The relationship obtained between  $\sigma^{\circ}$  and the *rms height* 193  
 independent of row direction, correlation length, and soil mois- 194  
 ture could be written as an exponential relationship of the form 195  
 $\sigma_{\text{dB}}^{\circ} = g(rms, \theta)_{\text{dB}} = \mu e^{-krms} + c$  [15], [16] or a logarithmic 196  
 relationship of the form  $\sigma_{\text{dB}}^{\circ} = g(rms, \theta)_{\text{dB}} = \mu \ln(rms) + 197$   
 $c$  [1]. 198

199 With taking into account of both soil roughness and soil 199  
 moisture, the radar signal in decibel scale may be written as 200

AQ8

AQ9

AQ10



TABLE II  
 INVERSION MODELS FOR ESTIMATING SOIL MOISTURE AND STATISTICS ON THE VALIDATION OF THESE MODELS

TerraSAR-X data - HH	Calibration phase Model	R <sup>2</sup>	Validation phase		
			Bias	std	RMSE
26°-28°	$mv(\%) = 2.31 \sigma_{dB} + 37.19$	0.79	0.52	2.76	2.81
50°-52°	$mv(\%) = 3.43 \sigma_{dB} + 54.30$	0.66	2.95	2.83	4.09
26°-28° and 50°-52°	$mv(\%) = 1.67 \sigma_{dB}(\theta_{low}) + 0.55 \sigma_{dB}(\theta_{high}) + 38.22$	0.69	1.65	2.46	2.91

201 the sum of two functions that describe the dependence of the  
 202 radar signal on soil moisture ( $f$ : linear) and surface roughness  
 203 ( $g$ : exponential) (e.g., [1] and [4])

$$\sigma_{dB}^{\circ} = f(mv, \theta)_{dB} + g(rms, \theta)_{dB} = \delta, mv + \mu, e^{-krms} + \tau \quad (3)$$

204 where  $k$  is the radar wavenumber ( $\sim 2 \text{ cm}^{-1}$  for TerraSAR-X).  
 205 This equation neglects the effect of the correlation length  
 206  $L$  on the backscattering coefficient. To take account of the  
 207 correlation length, Zribi and Deschambre [1] proposed a new  
 208 roughness parameter  $Z_s$ , defined by  $rms^2/L$ , which is the  
 209 product of the  $rms$  surface height and the slope of the soil  
 210 surface ( $rms/L$ ). Thus, the empirical model linking  $\sigma^{\circ}$  and  $Z_s$   
 211 could be written as  $\sigma_{dB}^{\circ} = \delta mv + \eta e^{-kZ_s} + \psi$ .

212 In the case of one SAR image characterized by one inci-  
 213 dence ( $\theta = 26^{\circ}-28^{\circ}$  or  $50^{\circ}-52^{\circ}$ ), inversion model is written as  
 214 follows:

$$mv = \alpha \sigma^{\circ}(\theta) + \beta. \quad (4)$$

215 The use of two incidence angles eliminates the effects of  
 216 roughness and thus allows linking the backscattering coefficient  
 217 to the soil moisture only. For two images acquired with low  
 218 and high incidence angles, the estimate of soil moisture can  
 219 be obtained by solving (3) for two incidences (substituting the  
 220  $e^{-krms}$  of  $\sigma^{\circ}(\theta_{low})$  into  $\sigma^{\circ}(\theta_{high})$ )

$$mv = \alpha \sigma^{\circ}(\theta_{low}) + \beta \sigma^{\circ}(\theta_{high}) + \gamma. \quad (5)$$

221  $\alpha$  and  $\beta$  depend on  $\delta$  and  $\mu$ , whereas  $\gamma$  is a function of  $\delta$ ,  $\mu$ ,  
 222 and  $\tau$  (in both incidence angles).

223 The form of (5) should be the same if the  $Z_s$  parameter was  
 224 used.

225 The empirical models given in (4) and (5) were then fitted to  
 226 experimental data acquired in 2008 and 2009 by using the least  
 227 squares method (cf. Table II). The validation of these models  
 228 was tested in using the data set of 2010 (13 points for each of  
 229 the two configurations HH26° and HH50°). The inputs are the  
 230 mean backscattering coefficients in decibels calculated for each  
 231 reference field.

#### 232 IV. RESULTS AND DISCUSSION

233 The inversion procedures were applied in order to retrieve  
 234 soil moisture. The results obtained in the validation phase  
 235 with one low incidence show inversion errors in the estimation

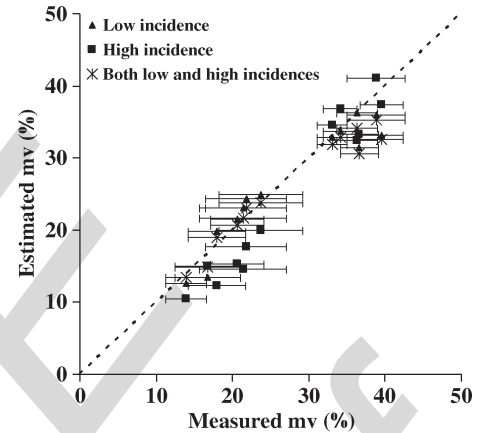


Fig. 4. Comparison between the estimated  $mv$  values and those measured. The error bars on the measured soil moisture values correspond to one standard deviation.

of  $mv$  of about 3% for incidence angles. The use of high  
 236 incidences ( $50^{\circ}-52^{\circ}$ ) gives slightly poorer results with an rmse  
 237 of about 4%. The accuracy of the soil moisture estimate remains  
 238 unchanged by using TerraSAR-X multi-incidence data (both  
 239 low and high incidence angles) with an rmse of about 3%  
 240 (Table II). Fig. 4 shows the good agreement between estimated  
 241 and measured  $mv$  values.

242 In contrast, large errors in the retrieved soil moisture were  
 243 observed at C-band for a single incidence angle (rmses of about  
 244 6% for  $20^{\circ}$  and 9% for  $40^{\circ}$ ) [4]. This is due to the fact that the  
 245 radar signal is much more sensitive to surface roughness at high  
 246 radar wavelength. The accuracy is strongly improved with the  
 247 use of both low and high incidences (rmse of about 3.5%) (e.g.,  
 248 [1], [2], and [4]).

249 The dependence of the radar signal at X-band on surface  
 250 roughness in agricultural areas was described as weak by  
 251 several works ([8], [14], and [17]). Results of these studies  
 252 show that the influence of surface roughness on the radar signal  
 253 increases with increasing radar wavelength. Moreover, this  
 254 dependence is mainly significant for low levels of roughness.  
 255 At X-band, Baghdadi *et al.* [4], [8] showed that the sensitivity  
 256 of  $\sigma^{\circ}$  to surface roughness becomes weak for  $rms > 1 \text{ cm}$ .  
 257 Thus, the effect of surface roughness on radar signal becomes  
 258 weak in X-band, which improves the estimates of soil moisture,  
 259 particularly for  $rms > 1 \text{ cm}$ . Moreover, the multi-incidence  
 260 approaches become less effective because the effect of surface  
 261 roughness that we try to eliminate is relatively weak at X-band  
 262 compared to C-band.

TABLE III  
 TERRASAR-X COVERAGE SIMULATION FOR ORGEVAL SITE BETWEEN  
 SEPTEMBER 2 AND 12, 2010 (ORBIT CYCLE)

Time	02 sep.	03 sep.	04 sep.	05 sep.	06 sep.	07 sep.	08 sep.	09 sep.	10 sep.	11 sep.	12 sep.
$\theta$ (°)	-	39	58	50	26	-	26	50	58	39	-

264

## V. CONCLUSION

265 This study examined the potential of TerraSAR-X data for  
 266 estimating soil moisture ( $mv$ ) over bare soils. TerraSAR-X  
 267 images collected between 2008 and 2010 over two study sites in  
 268 France were used. SAR images were acquired at HH polariza-  
 269 tion and for incidence angles of  $26^\circ$ ,  $28^\circ$ ,  $50^\circ$ , and  $52^\circ$ . The goal  
 270 of this work was to compare estimates of  $mv$  obtained from  
 271 various incidence configurations and to find the best sensor  
 272 configuration in incidence angle for measuring the bare soil  
 273 moisture.

274 This study tested empirical models for soil moisture inver-  
 275 sion from one incidence (low or high) and multi-incidence  
 276 TerraSAR-X data (both low and high incidences). The results  
 277 of this study may be summarized as follows.

278 1) For a single incidence, the retrieval algorithm performed  
 279 very well for low and high incidence angles. The rmses  
 280 for the soil moisture estimate are about 3% for  $26^\circ$ – $28^\circ$   
 281 and 4% for  $50^\circ$ – $52^\circ$ .

282 2) The accuracy of the soil moisture estimate does not  
 283 improve when two incidence angles (rmse is about 3%)  
 284 are used.

285 These results appear promising for the development of sim-  
 286 plified algorithms for retrieving soil moisture from TerraSAR-  
 287 X data and for monitoring temporal moisture changes. Table III  
 288 lists the different observation possibilities for the Orgeval study  
 289 site within one orbit cycle (11 days). This site could be imaged 8  
 290 times within 11 days (two images for each following incidence:  
 291  $\sim 26^\circ$ ,  $39^\circ$ ,  $50^\circ$ , and  $58^\circ$ ) and 24 times within one month.  
 292 The soil moisture mapping frequency with low incidence angle  
 293 ( $26^\circ$ ) or with both low and high incidence angles ( $26^\circ$  and  $50^\circ$ )  
 294 is possible six times within one month. The incidence of  $39^\circ$  can  
 295 also be used, which would increase to 12 the TerraSAR-X scene  
 296 number within one month. This very short revisit time makes  
 297 TerraSAR-X a very useful source for the soil moisture mapping.  
 298 Moreover, the increase in the acquisition frequency is much  
 299 awaited for the soil moisture data assimilation in hydrological  
 300 modeling.

301 In addition, the very high spatial resolution (metric) of the  
 302 TerraSAR-X sensor is also very promising for local estimation  
 303 of soil moisture at the within agricultural field scale. It offers a  
 304 great potential in terms of improving the quality of soil moisture  
 305 mapping for catchment areas where the parcels are of small  
 306 size.

307

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AQ11

## AUTHOR QUERIES

### AUTHOR PLEASE ANSWER ALL QUERIES

AQ1 = “In” was changed to “by.” Please check if the original thought was retained.

AQ2 = Please provide the expanded form of the acronym “COSMO-SkyMed.”

AQ3 = Please provide the expanded form of the acronym “ORFEO.”

AQ4 = “French Space Study Center” was changed to “National Space Study Center.” Please check if appropriate.

AQ5 = Please provide the expanded form of the acronym “UMR TETIS.”

AQ6 = The acronyms “CESBIO” and “IRD” were defined as “Centre d’Etudes Spatiales de la BIOSphère” and “Institut de Recherche pour le Développement,” respectively. Please check if appropriate.

AQ7 = Please provide the expanded form of the acronym “PALSAR.”

AQ8 = All occurrences of “ $2.2e^{-16}$ ” were changed to “ $< 2.2 \times 10^{-16}$ .” Please check if appropriate.

AQ9 = Please provide the expanded forms of the acronyms “ERS,” “RADARSAT,” and “ASAR.”

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# Use of TerraSAR-X Data to Retrieve Soil Moisture Over Bare Soil Agricultural Fields

Nicolas Baghdadi, Maelle Aubert, and Mehrez Zribi

**Abstract**—The retrieval of the bare soil moisture content from TerraSAR-X data is discussed using empirical approaches. Two cases were evaluated: 1) one image at low or high incidence angle and 2) two images, one at low incidence and one at high incidence. This study shows by using three databases collected between 2008 and 2010 over two study sites in France (Orgeval and Villamblain) that TerraSAR-X is a good remote sensing tool for the retrieving of surface soil moisture with accuracy of about 3% (rmse). Moreover, the accuracy of the soil moisture estimate does not improve when two incidence angles ( $26^\circ$ – $28^\circ$  or  $50^\circ$ – $52^\circ$ ) are used instead of only one. When compared with the result obtained with a high incidence angle ( $50^\circ$ – $52^\circ$ ), the use of low incidence angle ( $26^\circ$ – $28^\circ$ ) does not enable a significant improvement in estimating soil moisture (about 1%).

**Index Terms**—Soil moisture, TerraSAR-X images.

## I. INTRODUCTION

RADAR SIGNAL is a function of soil moisture and surface roughness in the case of bare soil. The possibility of retrieving these soil parameters was little investigated from X-band synthetic aperture radar (SAR). However, many studies were carried out by using C-band radar data (e.g., [1]–[4]). With the launch of satellites using the X-band ( $\sim 9.6$  GHz), such as TerraSAR-X and COSMO-SkyMed, the use of X-band data to derive soil parameters became possible. A radar configuration that minimizes the effects of surface roughness is recommended for a better estimate of soil moisture when using only one incidence angle. The optimal radar incidences in C-band for the retrieval of soil moisture are smaller than  $35^\circ$  [4].

Soil moisture estimation from SAR images is carried out by using physical or statistical models. Physical approach consists in using a physical model, such as the integral equation model [5], to predict the radar backscattering coefficient from SAR and soil parameters (wavelength, polarization, incidence angle, surface roughness, and soil dielectric constant). Statistical models based on experimental measurements are also often used in soil moisture estimation. For bare soils, the increase of radar angle ( $\sigma^\circ$ ) is supposed to be linear with the volumetric soil



Fig. 1. Location of study sites. (1) Orgeval. (2) Villamblain.

moisture for values between 5% and 35% [6]. Moreover,  $\sigma^\circ$  increases with soil surface roughness and follows an exponential or logarithmic behavior (e.g., [4] and [7]).

Very few studies analyzed the sensitivity of TerraSAR-X data to bare soil surface parameters. Baghdadi *et al.* [8] have observed that the radar signal at X-band is slightly more sensitive to surface roughness at high incidence angle than at low incidence angle. The difference observed between radar signals reflected by the roughest and smoothest areas increases with the radar wavelength. Moreover, results showed that the sensitivity of radar signal to surface roughness is better with PALSAR in L-band than with TerraSAR-X in X-band and that the C- and X-bands are similar sensitivity results. In this letter, only *in situ* soil moisture measurements in very wet conditions between 25% and 40% are available. Results obtained showed that the backscattering coefficient at X-band is stable when the moisture content ranges between 25% and 35% and that it decreases beyond this threshold.

Aubert *et al.* [9] have showed that the sensitivity of the TerraSAR-X signal to soil moisture is very important at low and high incidence angles. In comparison to results published with C-band SAR data, this sensitivity of the radar signal to soil moisture is higher in X-band. The second important result concerns the potential of the fine spatial resolution of TerraSAR (1 m) in the detection of soil moisture variations at the within-plot scale. The spatial distribution of slaking crust could be detected when soil moisture variation is observed between soil crusted and soil without crust. Indeed, areas covered by slaking crust could have greater soil moisture and, consequently, a greater backscattering signal than soils without crust.

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TABLE I  
 CHARACTERISTICS OF TERRASAR-X IMAGES AND SUMMARY OF GROUND-TRUTH MEASUREMENTS (*mv*, *rms*, AND *L*)

Date dd-mm-yy	Site	Pol.-Inc.	Fields number	<i>mv</i> (%) (min;max)	<i>rms</i> (cm) (min;max)	<i>L</i> (cm) (min;max)
06-02-08	Villamblain	HH-52°	8	(27 ; 34)	(1.3 ; 3.1)	(4.5 ; 9.1)
07-02-08	Villamblain	HH-28°	8	(27 ; 34)	(1.3 ; 3.1)	(4.5 ; 9.1)
12-02-08	Orgeval	HH-50°	6	(31 ; 36)	(1.8 ; 3.3)	(5.0 ; 9.3)
13-02-08	Orgeval	HH-26°	6	(31 ; 35)	(1.8 ; 3.3)	(5.0 ; 9.3)
17-03-09	Orgeval	HH-26°	7	(25 ; 32)	(1.7 ; 2.3)	(4.8 ; 6.9)
18-03-09	Orgeval	HH-50°	7	(24 ; 30)	(1.7 ; 2.3)	(4.8 ; 6.9)
25-03-09	Orgeval	HH-50°	3	(28 ; 29)	(2.0 ; 2.7)	(4.8 ; 5.7)
26-03-09	Orgeval	HH-26°	3	(24 ; 31)	(2.0 ; 2.7)	(4.8 ; 5.7)
08-04-09	Orgeval	HH-26°	6	(17 ; 26)	(1.1 ; 2.1)	(3.7 ; 6.0)
09-04-09	Orgeval	HH-50°	6	(15 ; 26)	(1.1 ; 2.1)	(3.7 ; 6.0)
01-03-10	Orgeval	HH-50°	6	(33 ; 40)	(1.9 ; 2.9)	(5.9 ; 7.5)
02-03-10	Orgeval	HH-26°	6	(33 ; 37)	(1.9 ; 2.9)	(5.9 ; 7.5)
12-03-10	Orgeval	HH-50°	7	(13 ; 25)	(1.1 ; 2.6)	(4.6 ; 7.0)
13-03-10	Orgeval	HH-26°	7	(15 ; 22)	(1.1 ; 2.6)	(4.6 ; 7.0)

71 At least one research question remained open. It concerns  
 72 the precision of the soil moisture estimates in bare agricultural  
 73 soils. The objective of this study is to examine the potential of  
 74 TerraSAR-X data for retrieving volumetric soil moisture over  
 75 bare soils. This work evaluates if the use of two incidence  
 76 angles at X-band [one low (26°–28°) and one high (50°–52°)]  
 77 improves the accuracy of the estimate of surface soil moisture  
 78 in comparison to only one incidence (low or high). TerraSAR-X  
 79 sensor has the advantage to acquire on the same study site  
 80 image pairs at low and high incidence angles within one day.  
 81 The goal of this work is to compare the findings with C- and  
 82 X-band data. At C-band, several studies have shown that the  
 83 use of two incidence angles provides distinct improvement in  
 84 the soil moisture estimate, in comparison with results obtained  
 85 using a single incidence (e.g., [1], [2], and [4]). Moreover,  
 86 low incidence angle is better than the high incidence angle  
 87 for estimating soil moisture with C-band SAR data. This letter  
 88 investigates this research question.

During the period of February–April (our SAR acquisitions), 99  
 the main crops are wheat and colza. They cover approximately 100  
 50% of the agricultural area. The remaining surface corre- 101  
 sponds to plowed soils awaiting future cultivation (corn and 102  
 potato). 103

### B. TerraSAR-X Images 104

Fourteen TerraSAR-X images (X-band ~9.65 GHz) were 105  
 acquired during the years of 2008, 2009, and 2010 (Table I). 106  
 The radar data are available in HH polarization, with incidence 107  
 angles ( $\theta$ ) of 26°, 28°, 50°, and 52°. The imaging mode 108  
 used was spotlight with a pixel spacing of 1 m. Radiometric 109  
 calibration using multilook ground range detected TerraSAR-X 110  
 images was first carried out using the following equation [10]: 111

$$\sigma_i^\circ(\text{dB}) = \log_{10}(Ks \cdot DN_i^2 - NEBN) + 10 \log_{10}(\sin \theta_i). \quad (1)$$

This equation transforms the amplitude of backscattered sig- 112  
 nal for each pixel ( $DN_i$ ) into a backscattering coefficient ( $\sigma^\circ$ ) 113  
 in decibels.  $Ks$  is the calibration coefficient, and  $NEBN$  is 114  
 the noise equivalent beta naught. All TerraSAR-X images were 115  
 then georeferenced using GPS points with a root-mean-square 116  
 error of the control points of approximately one pixel (i.e., 1 m). 117  
 This coregistration error was overcome by removing two 118  
 boundary pixels from each training plot relative to the limits 119  
 defined by the GPS control points. The mean backscattering 120  
 coefficients were calculated from calibrated SAR images by 121  
 averaging the linear  $\sigma^\circ$  values of all pixels within reference 122  
 fields. 123

## II. STUDY AREA AND DATA SET

89

### 90 A. Study Site

91 Data were acquired over two mainly agricultural sites  
 92 (Fig. 1). The Villamblain site is located in the south of Paris,  
 93 France (latitude 48°01' N and longitude 1°35' E) with soil  
 94 composed of 30% clay, 60% silt, and 10% sand. The second  
 95 site is situated in the Orgeval watershed, located in the east of  
 96 Paris, France (latitude 48°51' N and longitude 3°07' E). The soil  
 97 has a loamy texture, composed of 78% silt, 17% clay, and 5%  
 98 sand. Both of these two sites are very flat.

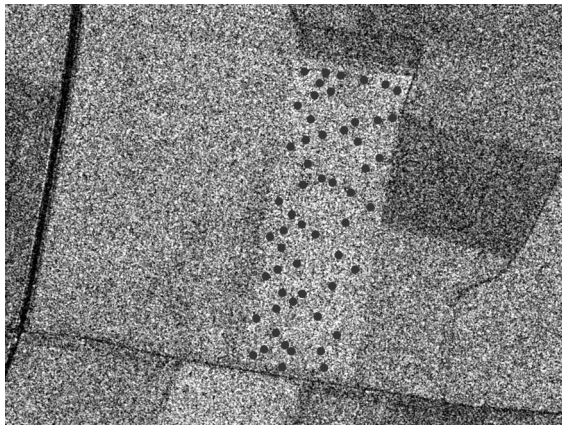


Fig. 2. Example of volumetric soil moisture measurements taken on a reference field.

### 124 C. Field Data

125 Simultaneously with TerraSAR-X acquisition, field mea-  
 126 surements of soil moisture and surface roughness have been  
 127 achieved on several bare soil reference fields of at least 2 ha.  
 128 In the case of TerraSAR-X in spotlight mode (pixel spacing of  
 129 1 m), this corresponds to a surface of 20 000 pixels or more.

130 The volumetric water content at field scale was assumed to be  
 131 equal to the mean value estimated from several samples (20–40  
 132 measurements per field; Fig. 2) collected from the top 5 cm  
 133 of soil using the gravimetric method. The soil moistures range  
 134 from 13% to 40%.

135 In most studies of microwave measurements carried out over  
 136 bare soils, the experimental relationship between soil moisture  
 137 and backscattering coefficient is provided by mean volumetric  
 138 water contents measured to a soil depth, generally 0–5 cm  
 139 or 0–10 cm. Indeed, only some studies using theory results  
 140 are available at X-band. These studies suggest a penetration  
 141 depth maybe lower than 5 cm. No experimental measurements  
 142 are made in field condition, and the low penetration depth  
 143 of X-band is only based on theoretical study. Therefore, the  
 144 penetration depth of the X-band is not yet well known.

145 Roughness measurements were made using needle pro-  
 146 filometers (1 m long and with 2-cm sampling intervals). Ten  
 147 roughness profiles were sampled for each training field (parallel  
 148 and perpendicular to the row direction). From these measure-  
 149 ments, the two roughness parameters, i.e., root mean square  
 150 (*rms*) surface height and correlation length (*L*), were calcu-  
 151 lated using the mean of all correlation functions. The *rms*  
 152 surface heights range from 1.1 to 3.3 cm, and the correlation  
 153 length (*L*) varies from 2.3 cm in sown fields to 9.3 cm in plowed  
 154 fields.

### 155 III. METHODOLOGY

156 The retrieval of soil moisture from TerraSAR-X images  
 157 by means of empirical approaches requires the development  
 158 of experimental relationships between  $\sigma_{\text{TerraSAR-X}}^{\circ}$  and the  
 159 measured soil moisture. TerraSAR data acquired in two config-  
 160 urations of incidence angles ( $\sim 26^{\circ}$  and  $\sim 50^{\circ}$ ) were used with  
 161 ground measurements conducted over bare soil. The sensitivity  
 162 of TerraSAR signal to soil moisture is the greatest for low

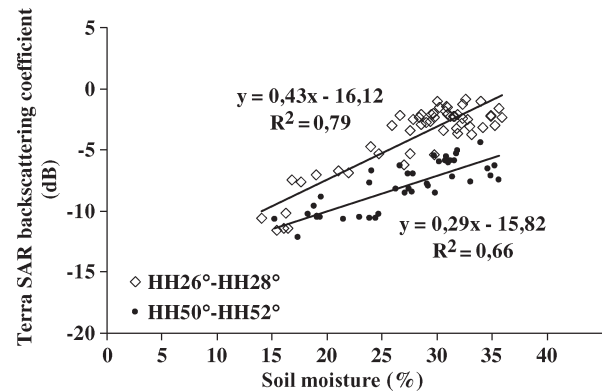


Fig. 3. TerraSAR-X signal versus volumetric soil moisture (measured at a depth of 5 cm). Each point corresponds to the average backscattering coefficient in decibels for one reference field. Thirty points are used for each of the two configurations HH26°–28° and HH50°–52° (data sets of 2008 and 2009).

incidence angle (0.43 dB/% for 26°–28° and 0.29 dB/% for 163  
 50°–52°; Fig. 3). For a confidence level of 95%, there are sig- 164  
 nificant relationships between the TerraSAR-X backscattering 165  
 coefficient and the *in situ* soil moisture because the *p*-values are 166  
 much less than 0.05 (*p-value* <  $2.2 \times 10^{-16}$  for HH26°–28° 167  
 and *p-value* =  $1.52 \times 10^{-10}$  for HH50°–52°). 168

169 Studies using C-band (ERS, RADARSAT, ASAR, etc.) 169  
 showed lower sensitivities between radar signal and soil mois- 170  
 ture, between 0.2 and 0.3 dB/% for low incidence angles 171  
 and about 0.1 dB/% for high incidence angles (e.g., [2] and 172  
 [11]–[13]). 173

174 The objective of this study is to analyze the influence of 174  
 incidence angle on the accuracy of the soil moisture estimate. 175  
 Configurations in HH polarization with single incidence an- 176  
 gle (26°–28° or 50°–52°) were studied. Next, multi-incidence 177  
 TerraSAR-X images acquired at both low and high  $\theta$  values 178  
 with one-day-spaced dates and only minor variations in soil 179  
 characteristics were used to analyze the possible improvement 180  
 in the soil moisture estimates when two incidences are used. 181

182 The empirical relationship between the radar backscattering 182  
 coefficient ( $\sigma^{\circ}$ ) and the volumetric soil moisture (*mv*) for bare 183  
 soil surfaces without taking into account the *rms* surface height 184  
 is given by (e.g., [14]; Fig. 3) 185

$$\sigma_{\text{dB}}^{\circ} = f(mv, \theta)_{\text{dB}} = \delta mv + \xi. \quad (2)$$

186 This simplified relationship is valid for *mv* values between 186  
 5% and 35% [6]. The coefficient  $\delta$  is dependent on SAR pa- 187  
 rameters (radar wavelength, incidence angle, and polarization), 188  
 while the coefficient  $\xi$  is controlled by SAR parameters and 189  
 surface roughness. Experimental data of  $\sigma^{\circ}$  and *mv* show slope 190  
 $\delta$  values of about 0.43 dB/% for HH26°–28° and 0.29 dB/% for 191  
 HH50°–52°. 192

193 The relationship obtained between  $\sigma^{\circ}$  and the *rms height* 193  
 independent of row direction, correlation length, and soil mois- 194  
 ture could be written as an exponential relationship of the form 195  
 $\sigma_{\text{dB}}^{\circ} = g(rms, \theta)_{\text{dB}} = \mu e^{-krms} + c$  [15], [16] or a logarithmic 196  
 relationship of the form  $\sigma_{\text{dB}}^{\circ} = g(rms, \theta)_{\text{dB}} = \mu \ln(rms) + 197$   
 $c$  [1]. 198

199 With taking into account of both soil roughness and soil 199  
 moisture, the radar signal in decibel scale may be written as 200

TABLE II  
 INVERSION MODELS FOR ESTIMATING SOIL MOISTURE AND STATISTICS ON THE VALIDATION OF THESE MODELS

TerraSAR-X data - HH	Calibration phase Model	R <sup>2</sup>	Validation phase		
			Bias	std	RMSE
26°-28°	$mv(\%) = 2.31 \sigma_{dB} + 37.19$	0.79	0.52	2.76	2.81
50°-52°	$mv(\%) = 3.43 \sigma_{dB} + 54.30$	0.66	2.95	2.83	4.09
26°-28° and 50°-52°	$mv(\%) = 1.67 \sigma_{dB}(\theta_{low}) + 0.55 \sigma_{dB}(\theta_{high}) + 38.22$	0.69	1.65	2.46	2.91

201 the sum of two functions that describe the dependence of the  
 202 radar signal on soil moisture ( $f$ : linear) and surface roughness  
 203 ( $g$ : exponential) (e.g., [1] and [4])

$$\sigma_{dB}^{\circ} = f(mv, \theta)_{dB} + g(rms, \theta)_{dB} = \delta, mv + \mu, e^{-krms} + \tau \quad (3)$$

204 where  $k$  is the radar wavenumber ( $\sim 2 \text{ cm}^{-1}$  for TerraSAR-X).  
 205 This equation neglects the effect of the correlation length  
 206  $L$  on the backscattering coefficient. To take account of the  
 207 correlation length, Zribi and Deschambre [1] proposed a new  
 208 roughness parameter  $Z_s$ , defined by  $rms^2/L$ , which is the  
 209 product of the  $rms$  surface height and the slope of the soil  
 210 surface ( $rms/L$ ). Thus, the empirical model linking  $\sigma^{\circ}$  and  $Z_s$   
 211 could be written as  $\sigma_{dB}^{\circ} = \delta mv + \eta e^{-kZ_s} + \psi$ .

212 In the case of one SAR image characterized by one inci-  
 213 dence ( $\theta = 26^{\circ}-28^{\circ}$  or  $50^{\circ}-52^{\circ}$ ), inversion model is written as  
 214 follows:

$$mv = \alpha \sigma^{\circ}(\theta) + \beta. \quad (4)$$

215 The use of two incidence angles eliminates the effects of  
 216 roughness and thus allows linking the backscattering coefficient  
 217 to the soil moisture only. For two images acquired with low  
 218 and high incidence angles, the estimate of soil moisture can  
 219 be obtained by solving (3) for two incidences (substituting the  
 220  $e^{-krms}$  of  $\sigma^{\circ}(\theta_{low})$  into  $\sigma^{\circ}(\theta_{high})$ )

$$mv = \alpha \sigma^{\circ}(\theta_{low}) + \beta \sigma^{\circ}(\theta_{high}) + \gamma. \quad (5)$$

221  $\alpha$  and  $\beta$  depend on  $\delta$  and  $\mu$ , whereas  $\gamma$  is a function of  $\delta$ ,  $\mu$ ,  
 222 and  $\tau$  (in both incidence angles).

223 The form of (5) should be the same if the  $Z_s$  parameter was  
 224 used.

225 The empirical models given in (4) and (5) were then fitted to  
 226 experimental data acquired in 2008 and 2009 by using the least  
 227 squares method (cf. Table II). The validation of these models  
 228 was tested in using the data set of 2010 (13 points for each of  
 229 the two configurations HH26° and HH50°). The inputs are the  
 230 mean backscattering coefficients in decibels calculated for each  
 231 reference field.

#### 232 IV. RESULTS AND DISCUSSION

233 The inversion procedures were applied in order to retrieve  
 234 soil moisture. The results obtained in the validation phase  
 235 with one low incidence show inversion errors in the estimation

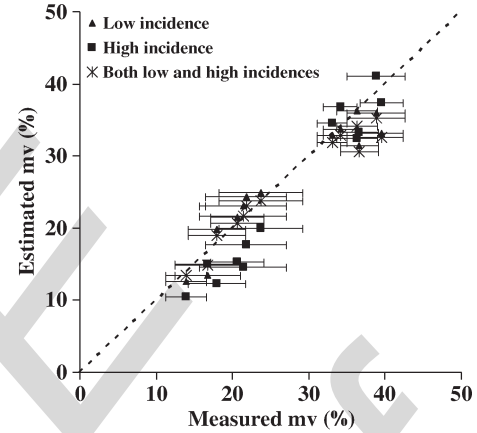


Fig. 4. Comparison between the estimated  $mv$  values and those measured. The error bars on the measured soil moisture values correspond to one standard deviation.

of  $mv$  of about 3% for incidence angles. The use of high 236  
 incidences ( $50^{\circ}-52^{\circ}$ ) gives slightly poorer results with an rmse 237  
 of about 4%. The accuracy of the soil moisture estimate remains 238  
 unchanged by using TerraSAR-X multi-incidence data (both 239  
 low and high incidence angles) with an rmse of about 3% 240  
 (Table II). Fig. 4 shows the good agreement between estimated 241  
 and measured  $mv$  values. 242

In contrast, large errors in the retrieved soil moisture were 243  
 observed at C-band for a single incidence angle (rmses of about 244  
 6% for  $20^{\circ}$  and 9% for  $40^{\circ}$ ) [4]. This is due to the fact that the 245  
 radar signal is much more sensitive to surface roughness at high 246  
 radar wavelength. The accuracy is strongly improved with the 247  
 use of both low and high incidences (rmse of about 3.5%) (e.g., 248  
 [1], [2], and [4]). 249

The dependence of the radar signal at X-band on surface 250  
 roughness in agricultural areas was described as weak by 251  
 several works ([8], [14], and [17]). Results of these studies 252  
 show that the influence of surface roughness on the radar signal 253  
 increases with increasing radar wavelength. Moreover, this 254  
 dependence is mainly significant for low levels of roughness. 255  
 At X-band, Baghdadi *et al.* [4], [8] showed that the sensitivity 256  
 of  $\sigma^{\circ}$  to surface roughness becomes weak for  $rms > 1 \text{ cm}$ . 257  
 Thus, the effect of surface roughness on radar signal becomes 258  
 weak in X-band, which improves the estimates of soil moisture, 259  
 particularly for  $rms > 1 \text{ cm}$ . Moreover, the multi-incidence 260  
 approaches become less effective because the effect of surface 261  
 roughness that we try to eliminate is relatively weak at X-band 262  
 compared to C-band. 263



TABLE III  
 TERRASAR-X COVERAGE SIMULATION FOR ORGEVAL SITE BETWEEN  
 SEPTEMBER 2 AND 12, 2010 (ORBIT CYCLE)

Time	02 sep.	03 sep.	04 sep.	05 sep.	06 sep.	07 sep.	08 sep.	09 sep.	10 sep.	11 sep.	12 sep.
$\theta$ (°)	-	39	58	50	26	-	26	50	58	39	-

264

## V. CONCLUSION

265 This study examined the potential of TerraSAR-X data for  
 266 estimating soil moisture ( $mv$ ) over bare soils. TerraSAR-X  
 267 images collected between 2008 and 2010 over two study sites in  
 268 France were used. SAR images were acquired at HH polariza-  
 269 tion and for incidence angles of  $26^\circ$ ,  $28^\circ$ ,  $50^\circ$ , and  $52^\circ$ . The goal  
 270 of this work was to compare estimates of  $mv$  obtained from  
 271 various incidence configurations and to find the best sensor  
 272 configuration in incidence angle for measuring the bare soil  
 273 moisture.

274 This study tested empirical models for soil moisture inver-  
 275 sion from one incidence (low or high) and multi-incidence  
 276 TerraSAR-X data (both low and high incidences). The results  
 277 of this study may be summarized as follows.

278 1) For a single incidence, the retrieval algorithm performed  
 279 very well for low and high incidence angles. The rmses  
 280 for the soil moisture estimate are about 3% for  $26^\circ$ – $28^\circ$   
 281 and 4% for  $50^\circ$ – $52^\circ$ .

282 2) The accuracy of the soil moisture estimate does not  
 283 improve when two incidence angles (rmse is about 3%)  
 284 are used.

285 These results appear promising for the development of sim-  
 286 plified algorithms for retrieving soil moisture from TerraSAR-  
 287 X data and for monitoring temporal moisture changes. Table III  
 288 lists the different observation possibilities for the Orgeval study  
 289 site within one orbit cycle (11 days). This site could be imaged 8  
 290 times within 11 days (two images for each following incidence:  
 291  $\sim 26^\circ$ ,  $39^\circ$ ,  $50^\circ$ , and  $58^\circ$ ) and 24 times within one month.  
 292 The soil moisture mapping frequency with low incidence angle  
 293 ( $26^\circ$ ) or with both low and high incidence angles ( $26^\circ$  and  $50^\circ$ )  
 294 is possible six times within one month. The incidence of  $39^\circ$  can  
 295 also be used, which would increase to 12 the TerraSAR-X scene  
 296 number within one month. This very short revisit time makes  
 297 TerraSAR-X a very useful source for the soil moisture mapping.  
 298 Moreover, the increase in the acquisition frequency is much  
 299 awaited for the soil moisture data assimilation in hydrological  
 300 modeling.

301 In addition, the very high spatial resolution (metric) of the  
 302 TerraSAR-X sensor is also very promising for local estimation  
 303 of soil moisture at the within agricultural field scale. It offers a  
 304 great potential in terms of improving the quality of soil moisture  
 305 mapping for catchment areas where the parcels are of small  
 306 size.

307

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AQ11



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